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Resource efficient connectivity restoration algorithm for mobile sensor/actor networks

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Abstract

Maintaining inter-node connectivity is of a paramount concern in most applications of mobile sensor/actor networks because nodes have to report their data and coordinate their operations. Failure of a node may partition the inter-node network into disjoint segments, and may thus hinder data delivery and inter-node coordination. This article presents a novel resource efficient connectivity restoration algorithm (RECRA) that opts to repair severed connectivity while imposing minimal overhead on the nodes. To avoid overreacting to non-critical failure, RECRA identifies critical/non-critical nodes and only triggers the recovery when a critical node fails. The failure of a node is detected by its neighbors and a recovery procedure is executed based on their proximity, status (critical/non-critical), and transmission range. RECRA prefers to employ a non-critical node and moves it to the place of failed node in order to limit the impact on coverage, scope of the recovery, and strangle successive cascaded relocations. In case non-critical nodes in the neighborhood are not available, the neighbors restore connectivity by exploiting their partially utilized transmission power and repositioning closer to the failed node. RECRA is validated analytically and through extensive simulations. The simulation results confirm the effectiveness of RECRA for both dense and sparse network segments and its performance advantage over contemporary schemes found in the literature.

Keywords: Wireless sensor and actor networks, Fault tolerance, Connectivity restoration, Node relocation, Topology repair

1. Introduction

Recent advances in sensing, wireless technologies, and smart actuation capabilities have paved the way for the emergence of small-sized and battery-operated mobile sensor/actor devices. These devices may be employed within the wireless sensor networks (WSNs) [1] to extend their capabilities. The mobile nodes vary from smart sensors such as Micabot [2] and Robomote [3] to quite capable actors such as iRobot[®] 710 [4] and Aerial mapping helicopter [5] that can deliver a response to an event. Augmenting WSN with mobile sensors and actors has led to the emergence of mobile sensor networks (MSNs) [3,6] and wireless sensor and actor networks (WSANs) [7], respectively. The role of an actor (or set of actors) is application dependent and is determined based on the requirements, environment, and the capabilities of actors. For example, an actor can deactivate a

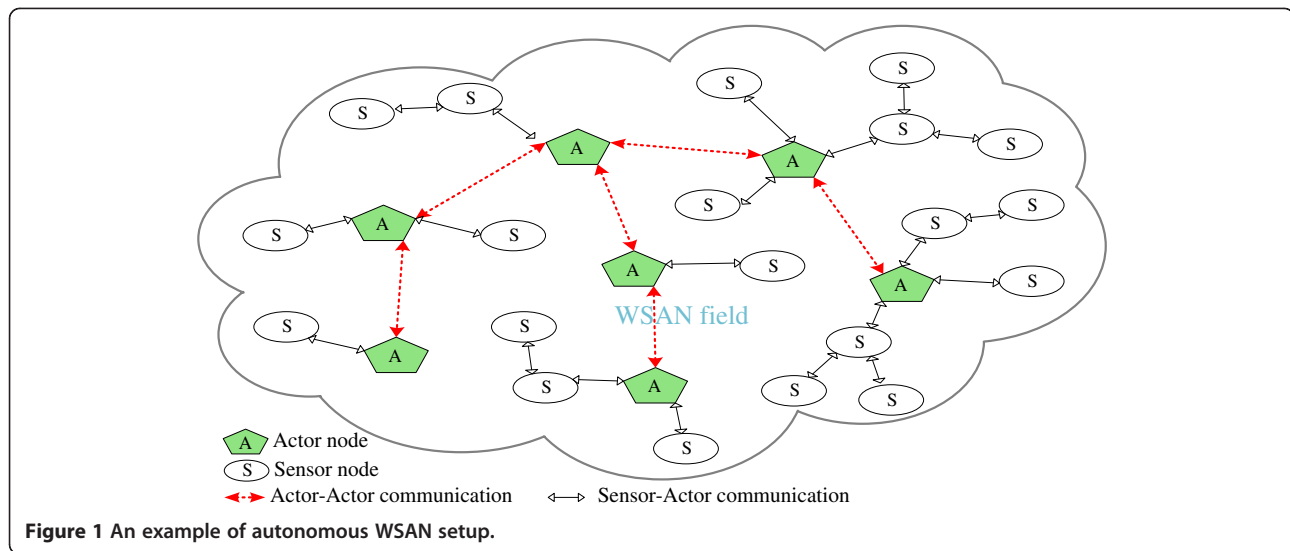
landmine, carry weapons, extinguish a fire, and rescue a trapped survivor. An articulation of a sample autonomous WSAN environment is shown in Figure 1.

In both MSNs and WSANs, maintaining network connectivity is extremely important throughout the network lifetime in order to satisfy application-level requirements. For example, effective and accurate monitoring of an area requires maximal presence of sensors to receive data from neighbors and perform aggregation or fusion. Similarly, most applications of WSAN necessitate availability of actors to establish and maintain a connected inter-actor topology in order to coordinate with each other on an optimal response and to synchronize their operations. For example, in disaster management applications, sensors detect and report the presence of survivors in the vicinity to the designated actors. The actors equipped with necessary equipment receive sensors data, process it, and share it with peer actors to determine the most appropriate set of actors that will participate in the operation. This requires that actors

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should be able to communicate with each other and perform appropriate action.

Nonetheless, the harsh application environment that MSNs and WSNs operate in makes nodes (i.e., sensors and actors) susceptible to physical damage and component malfunction. Failure of a critical node partitions the inter-node network into disjoint segments and consequently hinders data delivery and inter-node interaction. In addition, a node failure may cause significant loss of coverage. Since MSNs and WSNs operate autonomously in unattended setups, replacing the failed node is often infeasible and the recovery should be a self-healing and agile process that involves reconfiguring the inter-node topology in a distributed manner. Moreover, the resource constrained nature of MSNs and WSNs require the connectivity restoration process to be lightweight in terms of the incurred overhead and to limit the impact on network coverage.

This article presents a novel resource efficient connectivity restoration algorithm (RECRA) for repairing a topology after the failure of a critical node that caused partitioning the network topology into disjoint segments. In order to avoid unnecessary recovery overhead, RECRA classifies nodes into critical or non-critical to the network connectivity based on localized information. The neighbors detect the failure of a node F and decide to participate in the recovery based on their status, proximity, and how partially they currently utilize transmission power. To minimize the scope, impact, and overhead of recovery, RECRA prefers to employ non-critical nodes and moves them to the location of F . In case of unavailability of non-critical nodes, the neighbors of F collaboratively participate by gradually increasing their partially utilized transmission power and repositioning closer towards F . This is to minimize and balance recovery

overhead on individual nodes. The performance of RECRA is validated analytically and through extensive simulations. Simulation results confirm the effectiveness and efficiency of the proposed scheme compared to published schemes. To the best of the authors' knowledge, RECRA is the first reactive scheme that combines the use of non-critical nodes, node mobility, and partially utilized transmission power of nodes in the recovery process.

The rest of the article is organized as follows. The next section describes the system model and discusses the network partitioning problem. Section 3 summarizes the related work. The detail descriptions of the proposed RECRA algorithm, pseudocode, and analysis are provided in Section 4. The validation results and performance analysis of RECRA are presented in Section 5. Finally, Section 6 concludes the article.

2. System model and problem statement

RECRA is applicable to an MSN or a WSN. The mobile nodes can be a part of a flat network topology, in case of MSN, or form a second tier in hierarchical network architecture, e.g., actors or aggregation and forwarding units. The nodes are randomly placed in an area of interest where they discover each other and form a connected network. In the case of WSNs, this is an inter-actor network, i.e., no sensor nodes are involved. The nodes are assumed to be able to move on-demand with a consistent speed to reach their destination. However, this movement is assumed to be more costly than other operations such as computation and message transmission. The communication range r_{\max} of a node refers to the furthest Euclidean distance that its radio can reach. We assume that all nodes can dynamically adjust the output power of their radio when transmitting. This is very realistic assumption as popularly used radio

hardware such as CC1000 [8] and CC2420 [9] offer a register to specify the transmission power levels at run-time. Each node is assumed to maintain a list of its direct neighbors.

The effect of a node's failure depends on the position of that node in the network topology. For example, the loss of a leaf node, such as *N8* in Figure 2, has no negative impact on the inter-node reachability. Meanwhile, the failure of a critical, i.e., cut-vertex node such as *N1* partitions the network into disjoint segments. In order to tolerate the failure of a cut-vertex node, the existing schemes can be categorized into [10]: (i) proactive, (ii) reactive, and (iii) hybrid restoration. The proactive schemes provision fault tolerance by forming and maintaining a bi-connected topology. However, provisioning such a level of connectivity requires large node count, and thus boosts the cost and becomes impractical. On the other hand, in reactive schemes network responds only when a failure occurs. We argue that real-time restoration better suits MSN and WSANs since they are asynchronous and reactive in nature and it is difficult to predict the location and the scope of the failure beforehand. Hybrid schemes identify critical nodes in advance and designate them backup nodes as part of their contingency plan. The distributed and dynamic nature of MSN and WSANs makes hybrid schemes very suitable for mission-critical time-sensitive applications. RECRA falls in this category.

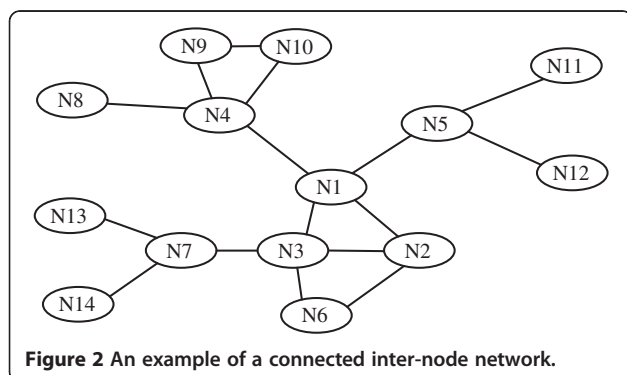
3. Related work

The issue of fault tolerance in context of MSNs and WSANs has also been studied in other contexts. For example, the fault-tolerant model presented in [11] designates multiple actors to each sensor and multiple sensors to each actor in order to guarantee event notification even in case of either failure or inaccessibility. However, our fault-tolerant model is in context of maintaining inter-node connectivity rather than reliable event notification delivery. Some researchers have proposed topology control algorithms to provide fault tolerance. They used to construct a fault-tolerant

topology by adjusting the node transmission power. For example, two distributed heuristics were proposed in [12] for homogeneous mobile networks to maintain connected topology by controlling the output power of the radios. The idea is to adjust the transmission power of nodes according to topology changes. Similarly, two localized algorithms for heterogeneous wireless networks were proposed in [13] to preserve a bi-connected network topology. Most of these schemes neither consider impact of node failure nor takes into account the limited communication range as bottleneck in maintaining connectivity.

Exploiting node mobility as a means of performance optimization has been pursued by multiple researchers both in the context of MSNs and WSANs. Energy conservation, increased connectivity and coverage, minimized latency and asset protection are the contemporary metrics targeted by the node repositioning. The reader is referred to [14] for a survey. Meanwhile, employing node mobility to repair damaged network topologies has only recently started to attract attention. The work can be categorized into block and cascaded movement. Block movement often requires a high pre-failure connectivity in order for the nodes to coordinate their response. An example of block movement-based approaches is the work of Basu and Redi [15], where the network is assumed to be 2-connected prior to the failure and the goal is to maintain such connectivity even under link or node failure. Their approach is centralized that may not be suitable for distributed and dynamic nature of MSANs. A distributed approach to tackle the same problem was presented in [16]. Unlike [15,16], the objective of this study is to maintain 1-connectivity. In the absence of higher level of connectivity, block movement is infeasible and nodes have to move in an independent manner, i.e., cascaded movement [6].

Approaches pursuing cascaded motion can be further categorized based on the network state that the individual nodes are assumed to maintain. Some approaches like DARA [17] and LDMR [18] base the node participation on having a list of 2-hop neighbors. DARA replaces the failed node "*F*" with one of its neighbor picked based on the node degree, distance, and ID, respectively. Repositioning the neighbor causes more links to break and the relocation process repeats in a cascaded manner. In LDMR, the neighbors move toward "*F*" and get replaced by nearest non-cut vertices. Basically, each neighbor broadcast a message looking for the best candidate that replaces it. On the other hand, some approaches such as RIM [19], VCR [20], and C³R [21] avoid the increased overhead for tracking 2-hop neighbors and require each node to be aware only of their 1-hop neighbors. The proposed RECRA algorithm fits in this category as well. RIM relocates the neighbors of "*F*" (irrespective of



whether the failed node and neighbors are critical/non-critical) inward until they become connected. Moving critical nodes inward requires children to follow and hence the scope of recovery may spread across the network. Albeit RIM balances the load and is suitable for sparse networks; however, its performance worsens significantly in dense networks. In addition, shrinking the topology inward causes significant loss of coverage.

Unlike RIM, the neighbors in VCR [20] volunteer by moving towards F while increasing their partially utilized transmission range until becomes connected. VCR applies a limited-scale diffusion in order to strangle repercussions of increased transmission power and improve coverage. However, VCR does not assess the impact of a node failure on connectivity and may engage critical nodes in recovery. Moreover, the diffusion in VCR is quite costly in terms of movement overhead especially in dense networks. Unlike RIM and VCR, RECRA identifies critical/non-critical nodes in advance and only triggers recovery in case of critical node failure. Moreover, RECRA prefers to employ non-critical nodes in recovery. Furthermore, it exploits the fact that some neighbors of " F " are not using their full communication range and would thus be able to reach more distant nodes than " F ". Consequently, RECRA employs fewer nodes and minimizes relocation overhead by exploiting their partially utilized transmission range.

A number of schemes cared for both connectivity and coverage. For example, C^3R employs node relocation in order to cope with the loss of coverage and connectivity when a node fails [21]. Instead of reconfiguring the network topology, neighbors move back and forth to replace the failed node in order to provide intermittent rather than permanent recovery. Obviously, this solution leads to frequent topology changes, imposes lots of overhead, and would thus become suitable as a temporary solution until spare actors are deployed. On the other hand, Akkaya and Janapala [22] address inter-actor connectivity and coverage at network setup time. Actors apply repelling forces to spread out and switch to attraction force when the actors become disconnected. However, they do not deal with actor failure.

4. Resource efficient connectivity restoration

After deployment, RECRA identifies critical/non-critical nodes in the network based on the localized information and neighbors monitor each other through heartbeats. The neighbors of a node F detect its failure through missing heartbeats and initiate a recovery process based on their status, proximity, and partially utilized transmission power. If there is a non-critical node, it will move to the position of F and announces the success of recovery. Otherwise, some of the critical neighbors pursue recovery by moving inward while increasing their

transmission power until becoming connected. RECRA is described in detail in the balance of this section.

4.1. Identifying critical nodes

As mentioned earlier, the failure of a critical node partitions the network into disjoint segments while the network stays strongly connected despite the loss of a non-critical node. Thus, it is imperative to assess the impact of a node failure on network connectivity in order to avoid unnecessary recovery overhead. In addition, involving critical nodes in the recovery process triggers cascaded node relocation and widens the scope of the recovery [10] which ultimately affect coverage. RECRA identifies critical/non-critical nodes in-advance in order to justify the need and minimize the overhead for tolerating a node failure. A number of algorithms have been proposed in the literature to determine cut-vertices in a graph. These algorithms can be categorized into centralized and distributed. In centralized algorithms [23,24], each node is required to maintain information about the entire topology which involves very high communication overhead, and thus may not be suitable for large-scale dynamic networks.

On the other hand, distributed and highly localized algorithms are more appropriate for large-scale MSNs and WSANs to support high mobility, scalability, robustness, and resource optimization. Localized algorithms can quickly determine critical/non-critical nodes with far less communication overhead as they only process localized information. This is extremely desirable in dynamic networks. However, these algorithms may classify some nodes as critical while they are not indeed critical in a network-wide view. With limited information, this is unavoidable because it is almost impossible for a node to know the alternate paths across the network. However, the accuracy of determining non-critical nodes is 100%. Considering the potential advantages of localized algorithms and the fact that RECRA prefers to engage non-critical nodes in recovery, such a category of approaches fits well and the reduced accuracy is not a major concern. Therefore, RECRA employs a localized cut-vertex detection procedure of [25] that only requires 1-hop position information and executes on each node in a distributed manner.

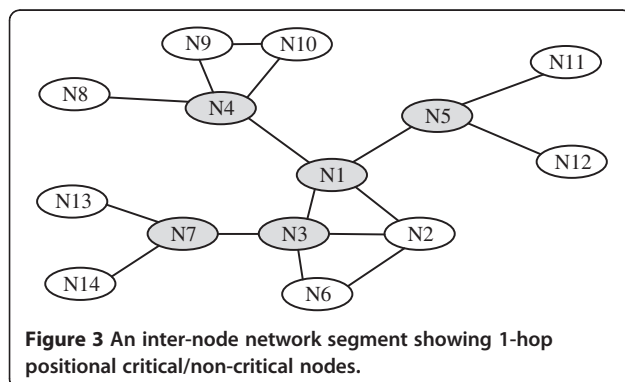
The number of critical/non-critical nodes in a network depends on the node density, communication range, and distribution of nodes. In general, there are two types of non-critical nodes, namely intermediate and leaf nodes. A good recovery scheme should exploit both types of non-critical nodes. The presence of intermediate non-critical nodes indicates a high degree of connectivity and overlapped coverage [10] and their population is typically high in dense networks. On the other hand, leaf nodes are located at the periphery or edges of the

network. Figure 3 shows an example topology. Nodes $N2$ through $N5$ are neighbors of $N1$ as can be observed. Node $N1$ is critical because its neighbors become disconnected without it. Meanwhile, node $N2$ is intermediate non-critical because its neighbors $N1$, $N3$, and $N6$ are connected without involving $N2$. Moreover, leaf nodes such as $N8$ are determined as non-critical, as their failure does not affect inter-node connectivity.

One may argue about the cost and implication of determining critical/non-critical nodes in RECRA. Each node in RECRA determines itself whether it is critical or not based on the localized information. This only involves computation overhead that is not a major concern compared to communication and movement overhead. Moreover, the distributed nature of RECRA equally divides this overhead on all nodes. Nevertheless, RECRA strives to minimize this computation overhead by executing this procedure only when a node or its neighbor changes its position. Second, the procedure is localized; therefore, it does not involve heavy computations. Distinguishing critical/non-critical nodes allows RECRA to avoid overreacting against non-critical node failures and preferably engage non-critical nodes in recovery which helps in minimizing recovery overhead.

4.2. Monitoring and failure detection

As part of their normal network operation, nodes periodically exchange heartbeat messages with their neighbors to update their status. The idea is based on monitoring vital signs of a body to determine the health condition. Similarly, nodes periodically update their status containing their ID, location, and criticality (i.e., critical/non-critical) to neighbors through heartbeat messages. The periodicity of heartbeat messages depends on many factors such as the mobility of nodes, the nature of the application, and the environment in which the network serves in. Missing a number of successive heartbeats from a node F is interpreted as an indication of its failure by the immediate neighbors. This number is based on the duty cycle for the data collection process and based on the susceptibility of nodes to failure. Therefore,



application designers are expected to assess these factors and determine the frequency and number accordingly. If the failed node F is non-critical there is no need to initiate any recovery as the network stays connected. On the other hand, if the failed node F is critical, the neighbors of F will apply RECRA and determine whether they will volunteer to participate in recovery or not based on the criteria described in the following section. Figure 4a shows that the failure of Node $N1$ divides the network into partitions and is detected by the neighbors.

4.3. Connectivity restoration

Upon detecting the failure of a critical node " F ", the neighbors initiate a recovery process. The scope of recovery mainly depends on the position of F and the status of the neighbors. In high-density networks, the probability of having non-critical nodes (intermediate and leaf) is more and moving these nodes will have minimal effect on inter-node connectivity and coverage. Therefore, RECRA prefers to exploit such an opportunity in the following manner:

- If there is a non-critical node in the neighborhood, it will move to the location of F and update its status to new neighbors, i.e., neighbors of F . Since the absence of a non-critical node (intermediate and leaf) does not inflict inter-actor connectivity, moving such a node to the place of F restores connectivity to the pre-failure level and does not require any further movement. Moreover, moving an intermediate non-critical node will almost preserve coverage loss as these nodes do have overlapped coverage with neighbors. Figure 4b shows that the non-critical node $N2$ moves to the location of the failed node and determines that it becomes critical at the new location. It updates its new status to neighbors to be ready for any future-related failure. Such an updated message also serves as a confirmation of the success in restoring connectivity.
- It might be possible that there is more than one non-critical node in the neighborhood of F . This situation is bit complicated especially when the nodes are unable to coordinate. If the non-critical nodes are neighbors, they can coordinate and with the node that has the highest node degree and is the closest to F replacing F . Repositioning such a node not only restores the connectivity, but also improves the coverage as shown in [10]. On the other hand, if non-critical nodes are unable to coordinate the recovery, they start moving towards F . The first one to reach F updates its status so that the remaining non-critical nodes stop and move back to their original location.

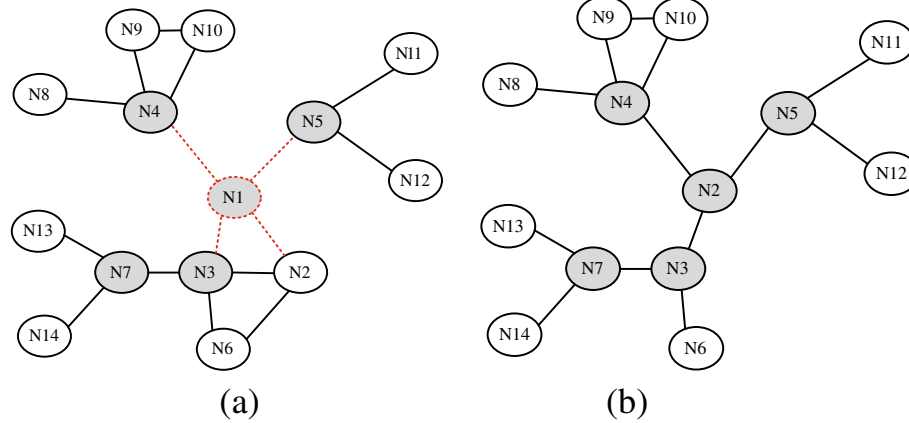


Figure 4 Failure detection and recovery when there is a non-critical node involved in recovery: (a) N2 detects the failure, (b) execute recovery process.

After failure detection, the critical neighbors of F wait for the non-critical neighbors to establish connectivity for a pre-defined period of time, i.e., the time it takes for a node to travel a distance r_{\max} . If the recovery message is received then the critical nodes are not required to participate in recovery. However, when the node density is low and connectivity is weak, the availability of immediate non-critical nodes in the neighborhood of F may be infrequent especially in the interior part of the network. Recovery in these situations is quite challenging and requires careful planning because moving a critical node indiscriminately may trigger a series of successive cascaded relocations that may prevail across the network. Such relocations impose high recovery overhead on individual nodes. Therefore, RECRA carefully orchestrates the recovery and collectively engages critical neighbors of F that meets the following criteria:

a. Participation criteria:

The critical neighbors of F perform their self-assessment in order to decide their role in recovery. This is based on their vicinity and how partially they utilized their transmission power.

- i. *Proximity*. A critical node $A \in \text{neighbors}(F)$ calculates its distance $d(A, F)$ to F . If $d(A, F)$ is more than a percentage α of its range r , node “A” is not required to participate in the recovery at this time, i.e., it stays as passive participant (PP). In other words, close neighbors to F are favored as active participants (APs) because a slight inward movement of neighbors around F collectively occupies the uncovered spot. Assuming uniform distribution of N nodes in a square area ($L \times L$), the distance between two nodes in the same row is $L/\sqrt{2N}$ and the distance between two diagonally neighboring nodes is $L/\sqrt{2N}$. Therefore,

the initial value of α is set as the average proximity to neighbors:

$$\alpha = .5 \left(\frac{L}{\sqrt{N}} + L \cdot \sqrt{2/N} \right)$$

It is worth noting that α is increased if node A is not connected within a preset time in order to recruit participants in the recovery. In other words, a PP can switch to AP depending on the progress in restoring restoration.

- ii. *Legibility factor (LF)*. The transmission power of nodes significantly affects the network connectivity. While power level at the transmitter determines the reachable range, i.e., how far the receiver can be, high power may increase interference and boost the number of exposed nodes [26] in dense networks. Therefore, power control is usually pursued in order to balance the interest in high connectivity and efficient utilization of the wireless channel. Particularly, nodes carefully set their transmission power to achieve signal-to-noise ratio that suits the intended receiver and limits the potential of medium access collision with other nodes in the vicinity. Moreover, power control is further employed in order to conserve energy. RECRA exploits the fact that many nodes are not utilizing their full range and would be able to boost their transmission power to reach further receiver than their neighbors. Thus, nodes whose range is partially utilized should be favored in the recovery process. The LF of a node captures the effect of the ratio of current range r_c to maximum range r_{\max} , i.e.,

$$LF = 1 - (r_c/r_{\max}),$$

Nodes with high LF are favored. A node will decide to become AP if its LF exceeds a preset threshold β . Initially, β can be approximated based on the node density. Assume the size of the deployment region is "Area". For a uniform node deployment, the value of r_c for establishing a connected network should be set such that

$$Area = N \cdot \pi \frac{1}{4} r_c^2,$$

where N is the number of nodes.

Using this equation, $r_c = \sqrt{4 \cdot Area / N \pi}$, which can be used to calculate an initial value of β . The initial β value is gradually decreased if no recovery is achieved within certain duration in order to increase the number of APs and restore connectivity.

b. Connectivity restoration

The APs exploit their partially unutilized transmission range and move towards F until they reach others. Connectivity restoration involves the following steps:

- i. *Node relocation.* The fact that a node is using a fraction of its maximum communication range r_{\max} indicates that this node can move away from its current spot and makes up for the increased proximity to its neighbors by boosting the output power of its radio. An AP node " P " would exploit this capability by moving to a distance γr_{\max} from F . Prior to departing its current position node " P " will notify its children. While moving, node " P " will increase its transmission power to stay connected to its children. If $d(P, F)$ exceeds $(r_{\max} - r_c)$, the children that are neither AP nor PP will follow to sustain their links to P , a step that referred in the literature as cascaded relocation [6]. RECRA opts to avoid or at least limits the scope of the cascaded relocation by favoring nodes that are close to F and

can grow their transmission power. At the start of the recovery process γ is set to 0.5. The rationale is that if all neighbors of F are at a distance $\frac{1}{2}r_{\max}$ away from F , the network becomes connected again [19].

- ii. *Connecting PP nodes.* The PP nodes will wait for APs to re-establish connectivity. The rationale is that APs will end up in the vicinity of F , yet not at the position of F . Therefore, there is a high probability for PP nodes to be able to reach one of those APs without a need to incur overhead. If a preset time passes without hearing from an AP, a PP will increase the value of α and/or and lowers β to become an AP. However, in this case an AP node will try to increase their transmission range first in order to find out whether other APs can be reached, before pursuing repositioning. When a PP becomes connected to an AP, it declares the success of the recovery (based on Theorem 1). For example, Figure 5a provides an illustration where the critical neighbors of $N2$ perform self-assessment based on the criteria prescribed earlier since there is non-critical neighbor to replace $N2$. Nodes $N4$ and $N5$ decide to become APs and move towards $N2$ while increasing their transmission power until establishing a link to PP, namely $N3$. Figure 5b shows that the recovery process not only connects the neighbors of $N2$, but also establishes some new links to $N4$ and $N7$. In addition, $N4$ and $N5$ maintain connectivity with their children, i.e., $N8, N9, N10$, and $N11, N12$, respectively. The children only adjust their transmission power to stay connected to their parents.

4.4. Pseudocode of RECRA

Figure 6 shows a high-level state diagram of RECRA algorithm. Upon detecting the failure of a neighbor F , the node switches from normal to recovery participant state. The transition from the recovery participant state would depend on the status, proximity, and LF. A non-critical

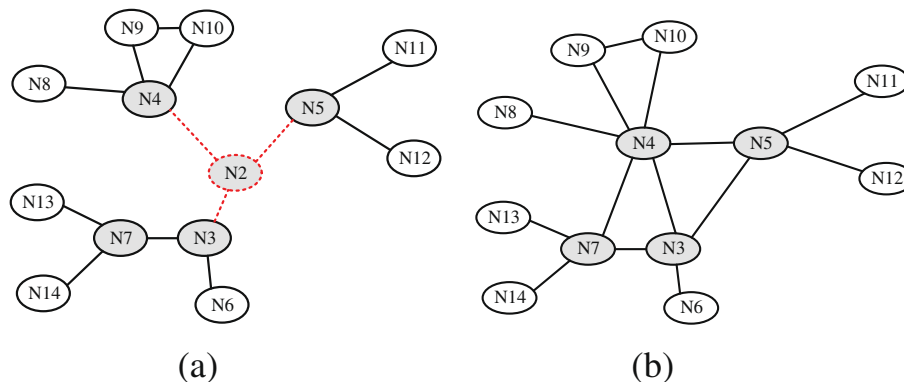


Figure 5 Failure detection and recovery in case of the unavailability of non-critical node.

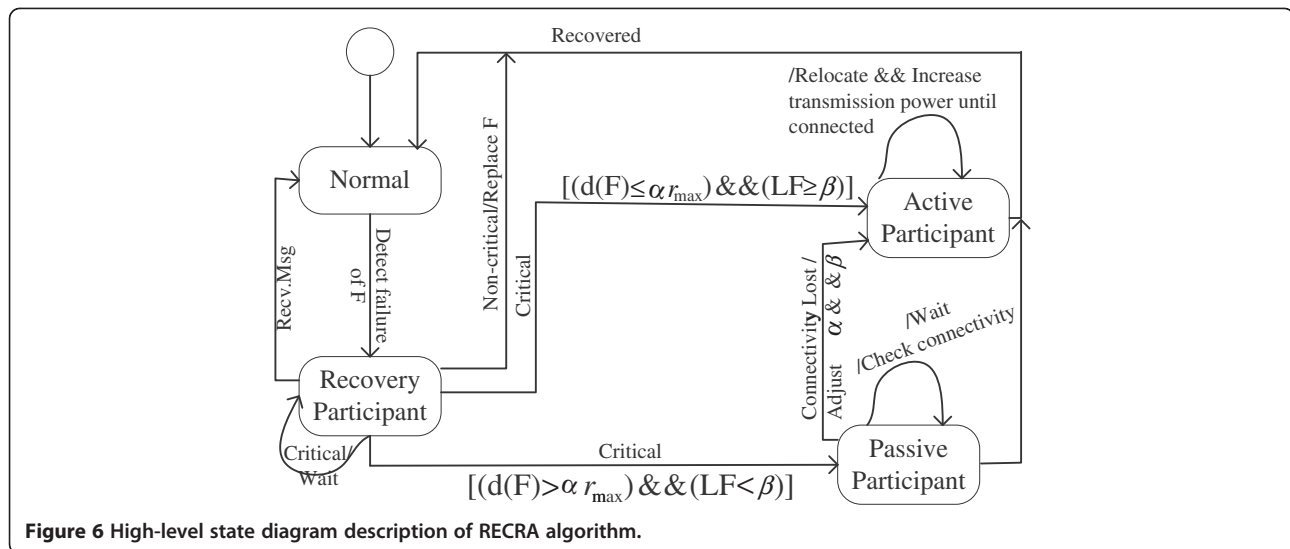


Figure 6 High-level state diagram description of RECRA algorithm.

node will simply replace the failed node and the recovery will be complete. Critical nodes will wait for the recovery message; if received they transition to a normal state and the recovery is complete. Otherwise, a critical recovery participant decides its role in the recovery based on its proximity and LF. A recovery participant node with close proximity to F and a high LF becomes an AP. Otherwise, the node transition to the PP state and keep tracking progress. In the AP state, a node performs relocation along with increasing transmission power until either it becomes connected with other APs or becomes at a distance γr_{\max} away from F . At the conclusion of the recovery the node switches back to the normal state. In the PP state, nodes wait for APs to re-establish connectivity. They continuously monitor the situation for a preset time. If the links are not established, the node times out and transitions to the AP state by increasing the value of α and/or and decreasing β . Again, in AP state it performs the same actions as described above.

Figure 7 shows the pseudocode of RECRA. The algorithm is to run on all nodes in a distributed manner. Initially, the status of all nodes is set to non-critical. A node A executes IdentifyCritical() procedure to determine whether it is critical or not based on the localized positional information (lines 1–5). Nodes periodically exchange heartbeat messages with neighbors to update status. If a node A did not receive consecutive heartbeats from neighbor node F , it marks the F as failed and sets a flag to indicate that recovery is incomplete (lines 6–12). This mimics the “recovery participant” state in Figure 5. Node A initiates a recovery process. If node A is non-critical, it looks for any other non-critical node that is a common neighbor with F . If there is another non-critical neighbor B with high node degree and least distance to F then B will move to the location of F . Otherwise,

A will replace F and send a recovered message (lines 13–18). The “while” loop (lines 20–39) reflects the actions in the AP and PP states. The loop repeats until the recovery is complete. Lines 22–29 are the steps taken in the AP state where children are notified and the actor moves towards F while increasing the transmission power. Otherwise, a PP state is declared in line 30 and the node monitors its connectivity to AP nodes. After waiting for τ time units, a PP node times out and adjusts α and β in order to switch to the AP state. The loop does not terminate until the node sets the “Recovered” indicator to true. The node may not be a neighbor of F and is rather a child to one of the AP nodes (lines 41–46). If node A receives a notification message from a parent, it watches its link to that parent. If connectivity is lost, the child will move towards its parent while increasing transmission power to re-establish the link.

4.5. RECRA algorithm analysis

This section proves the correctness and analyzes the performance of the proposed RECRA algorithm. We introduce the following theorems.

Theorem 1. *The network becomes strongly connected if it was strongly connected before a node F fails and if every PP node can reach an AP.*

Proof: If the network was strongly connected before F fails, every node should have a path to every other node in the network. The failure of F will affect the connectivity of the neighbors of F . Moving a non-critical node to the place of F will reestablish the connectivity among the neighbors of F as shown in Figure 8a. In case of unavailability of non-critical node, establishing links between those neighbors will make the network strongly connected again. When AP nodes move towards F while


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RECRA ()
IdentifyCritical ()
1  If  $\exists A \in N$  && Neighbors ( $A$ )  $\in N$ 
2      If (Neighbors ( $A$ ) become disconnected without  $A$ ) then
3           $A \rightarrow \text{IsCritical} = \text{True}$ 
4      Endif
5  Endif
FailureMonitoring ()
6  If ( $\exists A \in \text{Neighbors } (F) \in N$ )
7      Send/receive periodic heartbeat messages to update status
8      If ( $A$  did not receive successive heartbeats from  $F$ ) then
9           $F \rightarrow \text{Status} = \text{Failed}$ 
10          $A \rightarrow \text{Recovered} = \text{False}$ 
11      Endif
12  Endif
ConnectivityRestoration( $A, F$ )
13  If ( $A \rightarrow \text{IsCritical} == \text{False}$ ) then
14      If ( $\exists B \in \text{Neighbors } (A, F) \&\& B \rightarrow \text{IsCritical} == \text{False}$ ) then
15           $BC = \text{FindBestCandidate } (A, B)$ 
16           $\text{MoveToLocation}(BC, F)$ 
17      Endif
18       $\text{MoveToLocation}(A, F)$ 
19  Else
20      While (not Recovered)
21          Switch (True) {
22              Case: ( $d(A, F) \leq \alpha r_{\max}$ ) && ( $LF(A) \geq \beta$ )
23                  //  $A$  becomes an AP
24                  NotifyMove //notifies children before moving
25                  Move to a distance  $\gamma r_{\max}$  from  $F$ 
26                  Gradually increase transmission power
27                  If (connected to another AP) {
28                      Recovered = True
29                  } // End If
30              Case: ( $d(A, F) > \alpha r_{\max}$ ) && ( $LF(A) < \beta$ )
31                  //  $A$  becomes a PP
32                  Wait to hear from a AP within  $\tau$  time units
33                  If (message arrived from a AP  $P$ ) {
34                      Recovered = True
35                  }
36                  Else
37                      Reduce  $\alpha$  and/or  $\beta$ 
38                  } // End If
39              } // End switch
40          } // End while
41      } // End If
42      If (Notified by a AP) { // i.e.,  $A$  is a child node
43          If (lost connection to parent) {
44              NotifyMove // inform the children or  $A$ 
45              Move towards parent & increase transmission
46              power
47          } // End If
48      } // End If

```

Figure 7 Pseudocode for the RECRA algorithm.

increasing their communication range until become connected. Thus, if every PP can reach an AP node all neighbors of F will be connected again as shown in Figure 8c. In case neighbors of F are already using their maximum range then moving AP nodes to distance $r/2$ will connect them as shown in Figure 8b and already proven in RIM [19].

Theorem 2: In RECRA, the total movement distance is:

- $d_n \leq r_{\max}$, for a non-critical neighbor
- $d_{mc} \leq (r_{\max} - r) \sum_{i=0}^n C_i$, when nodes can increase their transmission power to r_{\max} and $r > \frac{1}{2}r_{\max}$
- $d_{cm} \leq \frac{r_{\max}}{2}(N - 1)$, For cascaded movement (assuming none of the neighbor can increase communication range).

where r_{\max} , n , and N are the communication range, the maximum number of neighbors, and the number of nodes in the network, respectively.

Proof: If any of the neighbors is non-critical, the failed node will simply be replaced Figure 8a. The worst-case scenario is when a non-critical node is located at distance r_{\max} from the failed node F . Hence, the total movement overhead will not exceed r_{\max} . The possibility of having non-critical nodes becomes higher with the increased network density and communication range. Therefore, the performance of RECRA is expected to improve in both the cases.

If there are no non-critical nodes in the neighborhood of F , nodes move towards F while increasing their communication range. The typical case, as articulated in Figure 8c, is when these neighbors are located at distance r from F and are also required to increase their communication range to r_{\max} to sustain connectivity with own neighbors. Therefore, in this case the total distance movement required to rejuvenate connectivity will not exceed $(r_{\max} - r) \sum_{i=0}^n C_i$.

The worst recovery scenario is when neighbors of F are critical nodes and are already using their maximum communication range. In this case, the neighbors move towards F by distance $\frac{1}{2}r_{\max}$ to become connected. The algorithm is recursively executed on the children of the moved nodes to become connected with parent. The maximum movement overhead will not be more than $\frac{r_{\max}}{2}(N - 1)$ when all healthy nodes in the network are critical and have to move as shown in Figure 8b.

Theorem 3: Participation of a non-critical from any partition or an AP from each partition is sufficient to restore connectivity irrespective of the number of partitions created.

Proof: There are two cases. In the first case, we may have two or more adjacent non-critical neighbors of F . Replacing F with a non-critical node will restore all the broken links and the recovery is complete as shown in Figure 9a. In the other case, participation of an AP from each partition not only restores connectivity, but also maintains existing communication links with neighbors. For example, Figure 9b illustrates that participation of a node B from partition#1 is not only sufficient to restore connectivity, but it also maintains existing links within the partition.

Theorem 4: RECRA impose maximum movement overhead of r_{\max} on each recovery participant, where r_{\max} is the maximum communication range of nodes.

Proof: Since a failed critical node is either replaced by one of its non-critical neighbor or the recovery participant among neighbors moves toward F while increasing their transmission power. In the former case, the non-

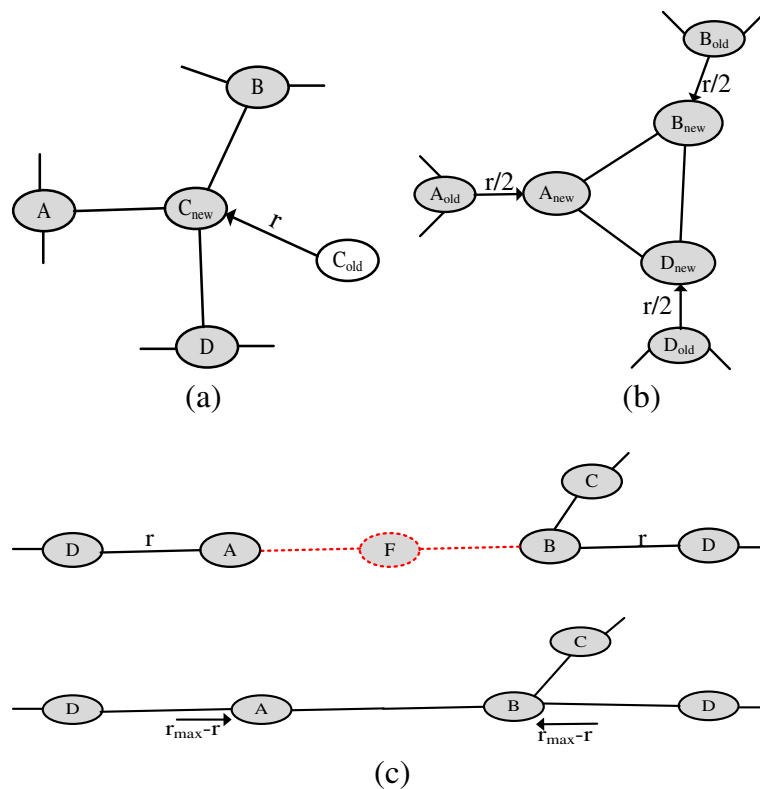


Figure 8 Illustrating recovery scenarios: (a) non-critical replacement, (b) moving critical inward, (c) failure detection and recovery based on movement and communication.

critical node travels a maximum distance of r_{\max} to substitute F in worst case. While in the latter case, each recovery participant may have to travel distance $r_{\max}/2$ in worst case as proven in Theorem 2. Therefore, the maximum movement overhead in RECRA is r_{\max} because each node moves only once.

Theorem 5: The time for RECRA to successfully restore connectivity is $O(f + T)$, where f is the fault detection latency and T is the maximum movement time of any node, which is proportional to r_{\max} .

Proof: In RECRA, the worst-case time complexity is proportional to the failure detection and recovery. Let us

assume that the time to detect a failure or receive a movement notification message is f . In the worst case, the time it takes to establish connectivity is equal to $T = \frac{r_{\max}}{2s}(N - 1)$ where s is the movement speed of nodes and the $\frac{r_{\max}}{2}(N - 1)$ is the longest distance travelled by nodes when cascaded relocation is pursued, as proved in Theorem 2. Since, the time to increase transmission power is nominal and it is pursued during relocation, therefore, it is included in T . Therefore, the convergence time of RECRA to restore connectivity in the worst case is $k(f + T)$, which is $O(f + T)$.

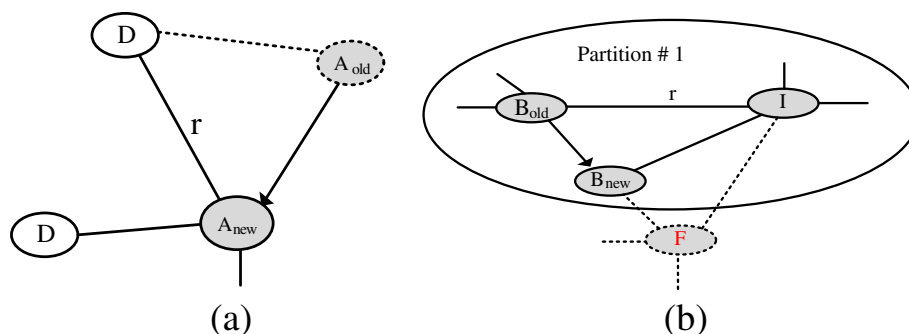


Figure 9 Demonstrating the sufficient condition for neighbor participation: (a) non-critical neighbors, (b) critical neighbors.

Theorem 6: *The message complexity of the RECRA is $O(N)$, where N is the number of nodes in the network.*

Proof: RECRA entails one message transmission for each node as it is a localized scheme that only maintains 1-hop neighbor information to restore connectivity. In addition, every AP (i.e., critical node participating in recovery) has to send one movement notification message to its children so that they can sustain connectivity with the parent and to avoid being wrongfully perceived as faulty. At the new location, the node will send one message to its neighbors. Therefore, in worst case, when all nodes move, the total number of messages will be $(N + 2|AP|)$. Therefore, the message complexity of RECRA is $O(N)$.

5. Experimental evaluation

The effectiveness of the RECRA is validated through extensive simulation experiments. This section describes the experiment setup, performance metrics, and experimental results.

5.1. Experiment setup and performance metrics

We have developed a customized simulation environment. In the experiments, we have created topologies that consist of varying number of nodes (20–100). Nodes are randomly placed in an area of $1000 \times 600 \text{ m}^2$. We have varied the transmission range of nodes (50–125) during experiments. Free space signal propagation model and collision-free medium access arbitration are assumed. The performance is assessed using the following metrics.

- The *total distance moved* by all nodes involved in the recovery: This gauges the efficiency of RECRA in terms of the motion-related overhead, e.g., dissipated energy during recovery.
- The *number of nodes* moved during the recovery: This metric reflects the scope of the recovery process the level of disturbance caused to the existing topology. This is crucial specifically in mission-critical applications.
- The *number of messages exchanged* among nodes: Again this metric indicates the recovery overhead.
- The *percentage of coverage change* (increase or decrease) relative to the pre-failure level: Although connectivity is the prime objective of RECRA, node coverage is important for many setups. The loss of a node usually has a negative impact on coverage. This metric assesses whether RECRA alleviates or worsens the coverage loss.
- The *degree of connectivity* in the network corresponding to pre-failure level: This metric reflects the level of connectivity that indicates the availability of node-independent paths. It is

measured by averaging the number of neighbors that a node has.

The following parameters were used to vary the network configuration in the experiments:

- The *number of deployed nodes* (N) in the network affects the node density and the inter-node connectivity.
- The *node communication range* (r) is the initial communication range and is equal for all nodes. Whereas, " r_{\max} " is the maximum communication range a node can have.
- The *transmission range utilization ratio* (p), which is defined as r/r_{\max} , affects the connectivity and recovery overhead.

5.2. Baseline approaches

We compare the performance of RECRA to that of DARA [17], RIM [19], and VCR [20] based on the aforementioned metrics. Like RECRA, all these approaches are distributed, reactive, and exploit node relocation in order to restore connectivity. However, their procedure is different. When a node F fails, DARA selects a best candidate A among its 1-hop neighbors and replaces it. The algorithm is recursively applied to tolerate connectivity loss due to movement, i.e., A will be replaced with one of its neighbors and so on. On the other hand, RIM moves all the 1-hop neighbors towards F until they become connected. RIM then applies cascaded relocation to re-establish the links that get severed by nodes movement. VCR exploits partially utilized transmission power of the 1-hop neighbors besides relocation. A controlled diffusion is applied among recovery participants to increase the coverage and limit the repercussions of recovery.

5.3. Simulation results and analysis

The simulation experiments involve randomly generated topologies with varying number of nodes, communication ranges, and their utilization ratios. The number of nodes has been set to 20, 40, 60, 80, and 100. The communication range " r " of nodes is changed among 50, 75, 100, and 125. The transmission range utilization ratio has been fluctuated to 10, 25, 50, and 75%. When changing the node count, " r " and " p " are fixed at 100 m and 50%, respectively; and " N " is set to 100 while varying the communication range and to 60 when changing " p " unless stated otherwise. The values of α and β are calculated based on N and r . Initially, the values of γ and τ are set to 0.5 and γr_{\max} , respectively. We carefully picked the failed node in these experiments to ensure that it is really a cut vertex since none of the baseline approaches determine critical/non-critical nodes in advance. In

addition, the prime objective of this study is to minimize the post-failure recovery. Moreover, we also ran experiments in which the failed node is randomly picked in order to assess the impact of inaccuracy of determining critical nodes using 1-hop (RECRA) and 2-hop (DARA) information. In these experiments, “ r ” and “ p ” are fixed at 100 m and 25%, respectively. The results of individual experiments are averaged over 20 trials to ensure statistically stable results. All results are subject to 90% confidence interval analysis and stays within 10% the sample mean.

5.3.1. Total traveled distance

The results for the total distance traveled by all nodes until connectivity is restored are shown in Figures 10 and 11, respectively. Figure 10a,b clearly indicates that the performance of RECRA is not much affected by the change in node density and communication range. In addition, RECRA consistently outperforms baseline approaches in terms of the total distance movement. This is because RECRA prefers to employ non-critical nodes in the recovery which does not necessitate cascaded relocations. Moreover, RECRA strives to limit the involvement of nodes that are far from the failed node and restrain the movement overhead by pursuing higher transmission range. The results demonstrate that RECRA is effective for both sparse and dense networks.

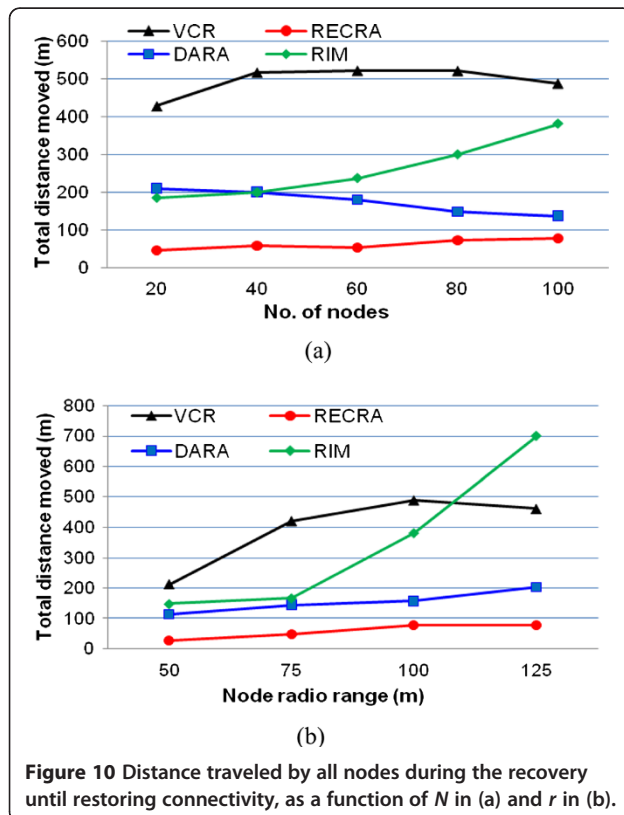


Figure 10 Distance traveled by all nodes during the recovery until restoring connectivity, as a function of N in (a) and r in (b).

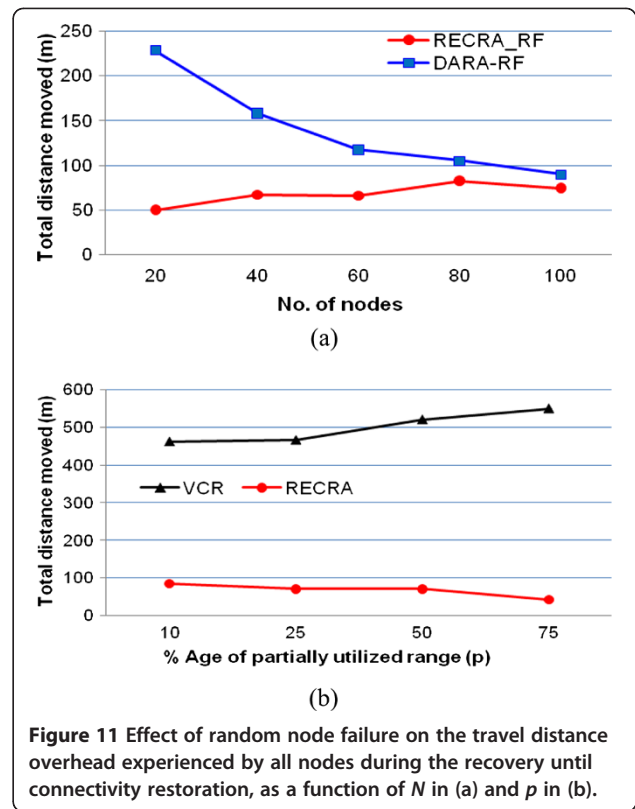


Figure 11 Effect of random node failure on the travel distance overhead experienced by all nodes during the recovery until connectivity restoration, as a function of N in (a) and p in (b).

At first glance, one would wonder why the movement overhead is not reduced for higher N and r . This is because RECRA may choose a distant non-critical node instead of moving nearby neighbors that can increase transmission power as is evident from Figure 10b.

Figure 10a indicates that the performance of RECRA scales very well and is not much affected by the node density given the optimized selection of nodes as explained in Section 4.3. In sparse networks, where a non-critical node is not available in the neighborhood, RECRA collectively engages all neighbors of F that can exploit their partially utilized transmission power. On the other hand, RECRA entirely relies on movement in case of a non-critical node in order to minimize scope of recovery as we shall discuss later in this section. Similar observation can be made for the communication range (Figure 10b), where the connectivity-restoration overhead is minor compared to the baseline approaches.

Figure 10 also shows that the performance of RIM and VCR is significantly affected when increasing N and r . This is attributed to the fact that the number of neighbors increases in both cases. The self-spreading step in VCR is costly in terms of the motion overhead. This is mainly because the scope of the motion is wider and involves nodes that do not have to relocate for restoring the connectivity. It is worth noting that self-spreading will boost the coverage achieved by VCR as shown later

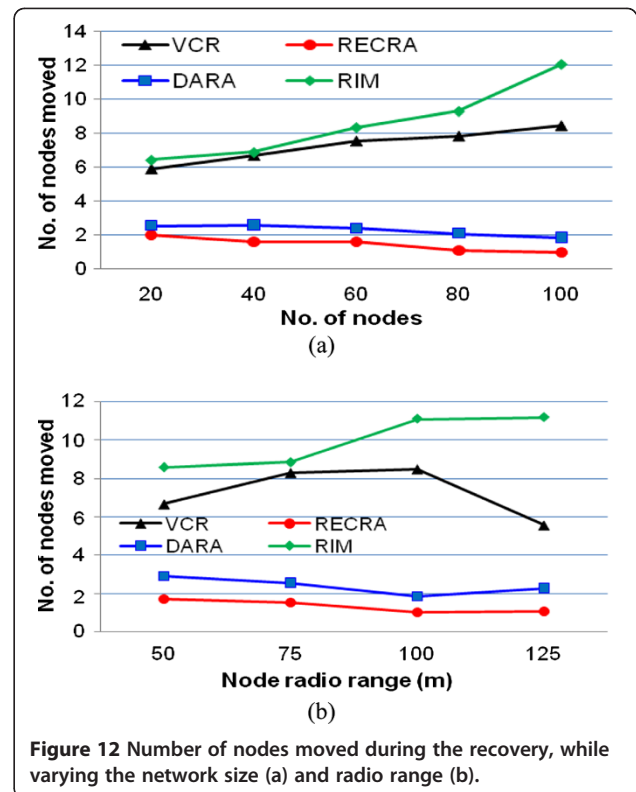
in this section. However, RECRA prefers to minimize movement-related overhead while keeping intact the pre-failure coverage. Figure 10 also indicates that the motion overhead in DARA is slightly more than RECRA. This is because DARA moves least degree nodes that are often critical and results in successive relocations. However, the higher density helps DARA to find leaf nodes and limits the scope of recovery.

Figure 11a reports the impact of critical node detection accuracy which can potentially worsen the performance by unnecessarily triggering recovery of uncritical node. In this experiment, the failed node is randomly picked and the criticality assessment is conducted by applying the algorithm of [27] for DARA and [25] for RECRA using 2-hop and 1-hop information, respectively. The performance results show that RECRA still imposes less movement overhead despite the fact that DARA can perform more accurate assessment of the criticality of the failed node, as pointed out in Figure 10a. In sparse networks, the accuracy of determining critical nodes with 1-hop and 2-hop is almost similar, since the node degree is low, and thus the performance stays consistent with Figure 10a. However, with high node density, the inaccuracy of critical node detection with 1-hop grows in significance compared to 2-hop due to the increased level of connectivity in the network. That is why the performance of DARA becomes quite closer to RECRA for network with high node count. On the other hand, RECRA may overreact in some cases but it strangles the repercussions through high availability of non-critical nodes.

Figure 11b shows the impact of the transmission range utilization ratio " p " on the movement overhead. Again, the performance of RECRA is almost consistent and far better than VCR. This is mainly because RECRA prefers to engage non-critical nodes and avoids the need to perform self-spreading. The performance of RECRA is slightly improved with the higher values of " p ". This is expected because nodes reduce their reliance on movement and exploit their partially utilized transmission range. On the other hand, the performance of VCR is negatively affected since nodes have to further move away during self-spreading.

5.3.2. Number of moved nodes

Figure 12 shows the number of recovery participants when RECRA and the baseline approaches are applied. The performance graphs confirm the advantage of RECRA which moves fewer nodes than contemporary schemes. This is because it limits the scope of recovery and avoids recursive cascaded relocations by employing non-critical nodes and exploiting partially utilized transmission ranges. Moreover, the performance of RECRA remains almost constant while varying the number of



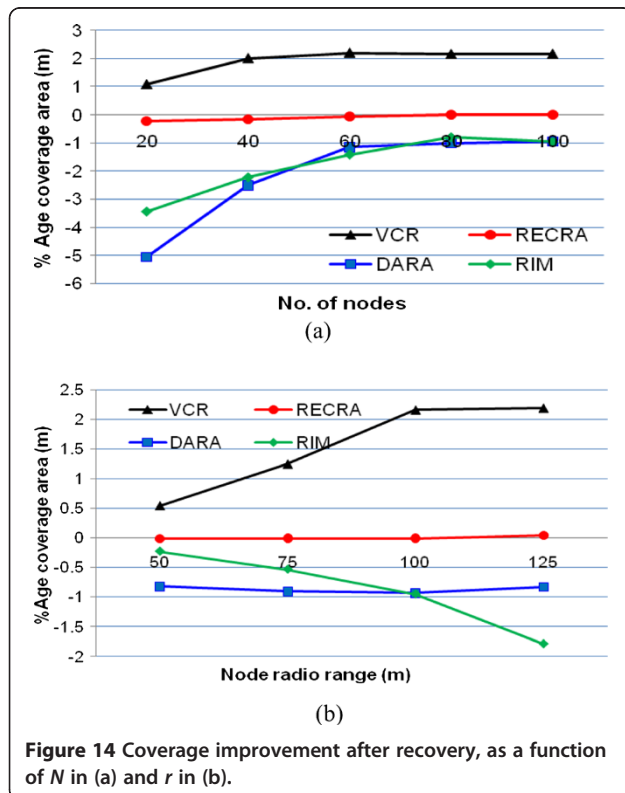
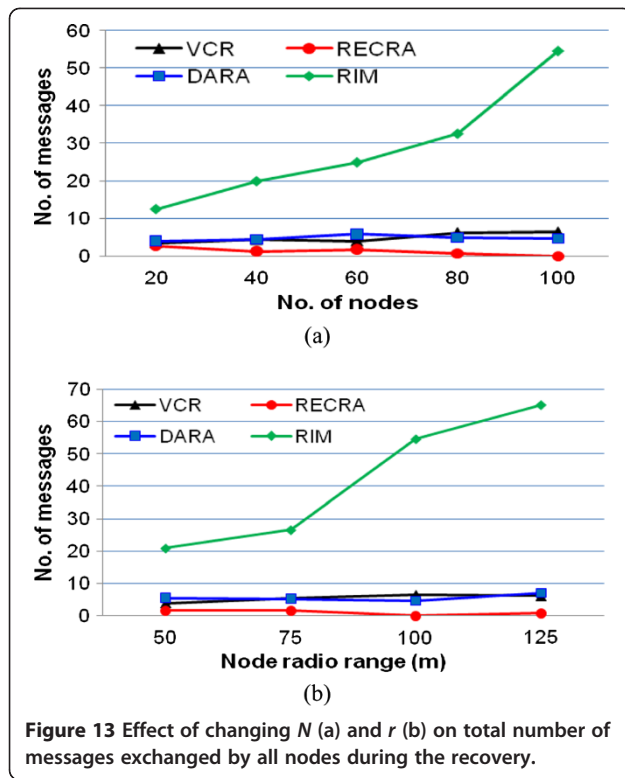
nodes and their radio range, which indicates great scalability.

5.3.3. Number of messages exchanged

Figure 13 reports on the messaging overhead as a function of the network size and radio range. As indicated in the figure, RECRA introduces less messaging overhead compared to the baseline approaches. This is because RECRA prefers to employ non-critical nodes in the recovery which does not require coordination messages. Moreover, RECRA strives to engage only the closest nodes among 1-hop neighbors of F . On the other hand, Figure 13 indicates that the messaging overhead in RIM significantly grows for a high actor density and long communication range because the number of neighbors increases in both cases.

5.3.4. Percentage of coverage improvement

Figure 14 shows the impact of changing N and r on coverage, measured in terms of percentage of improvement relative to the pre-failure level. The sensing/action range is set to 50 m in these experiments. Both figures indicate that RECRA almost preserves the pre-failure coverage despite the fact that the prime concern is resource efficiency in terms of recovery overhead. This result is attributed to the fact that neighbors do have overlap coverage with each other depending upon the node density and distribution. Moving an intermediate



non-critical node will only reduce the overlap coverage and occupy the spot vacated due to the failure of F . Moreover, constrained movement of neighbors around F inward will also help covers the bare spots. Figure 14a shows that high node density helps RECRA to preserve pre-failure coverage because of availability of intermediate non-critical nodes and overlapped coverage among nodes. However, RECRA struggles to sustain pre-failure coverage in sparse networks because of the aforementioned reason.

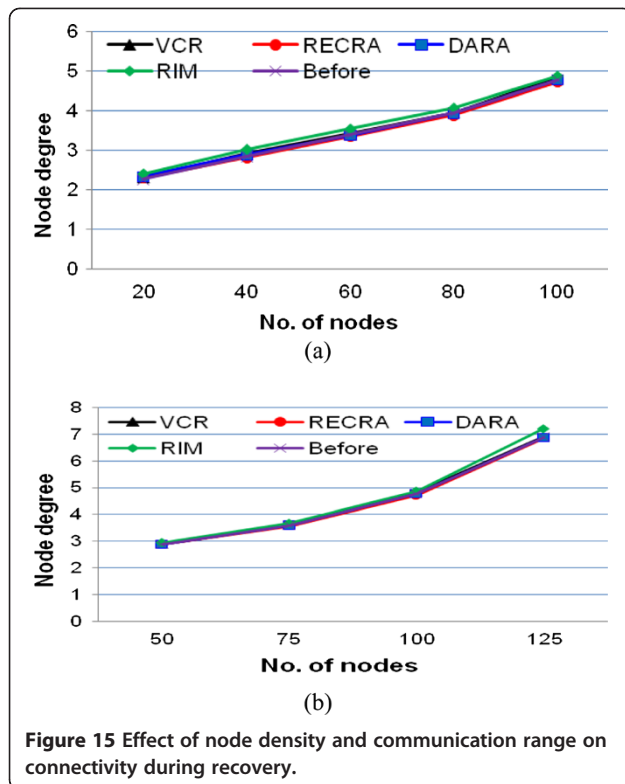
On the other hand, increasing the node density in Figure 14a helps DARA and RIM; yet they still do not make up for the coverage loss and definitely do not match RECRA's performance. DARA prefers to choose the least degree nodes. A node with the least degree is often critical and moving such a node only shifts the coverage hole from one place to another. In RIM, the topology gets shrunk inward which leads to significant loss of coverage at network periphery. Whereas, VCR improves coverage by about 2% and consistently outperforms both baseline schemes. This coverage advantage of VCR is mainly driven by the diffusion, spreading, of nodes that is performed after connectivity restoration in order to limit the effect of signal interference. Figure 14b indicates that for VCR the coverage grows with the increase in the communication range while the performance of DARA is not affected much. On the other hand, the performance of RIM significantly worsens when growing the communication range. With the increased value of r , the network becomes more connected and the number of neighbors of F grows. RIM moves nodes inwards making the area around F to be more crowded than at the network periphery and thus cause a significant loss of coverage.

5.3.5. Degree of connectivity

Figure 15 demonstrates the degree of connectivity maintained by all the approaches relative to the pre-failure level. Again, this metric reflects the average node degree in the network and is measured by averaging the number of neighbors for the individual nodes. Both figures clearly indicate that RECRA consistently maintains the same degree of connectivity as of other approaches, despite the fact that the prime objective of RECRA is resource efficient recovery. This is due to moving non-critical nodes, limiting the scope of relocation, and increasing transmission power. Figures 10, 11, 12, 13, and 14 confirm that RECRA strikes a balance between various objectives.

5.4. Important observations

We would like to make few important observations regarding the performance and utility of RECRA. Maintaining network state information at each node (i.e.,



1-hop, 2-hop, etc.) dominates the accuracy of determining cut-vertices. The fact that RECRA and DARA maintain 1-hop and 2-hop information, respectively; therefore, the accuracy of determining cut-vertices in DARA is slightly higher and may cause RECRA to overreact in some cases. However, in dynamic networks, maintaining more network state information at each node consistently requires extra overhead. Since, the prime focus of this study is resource efficient recovery; therefore, RECRA strives to avoid such unnecessary overhead and prefers to engage non-critical and/or nodes with partially utilized transmission range. The experimental results presented above have confirmed the superiority of RECRA over contemporary schemes.

The possibility of simultaneous node failure in MSNs and WSNs is rare; therefore, RECRA is designed to recover from a single node failure. In handling simultaneous failures may cause conflicting situations for RECRA where non-critical nodes in the neighborhood are not available.

6. Conclusion and future work

This article has presented a novel distributed resource-efficient connectivity restoration algorithm for MSNs and WSNs. Unlike most published schemes, RECRA exploits availability of non-critical nodes and partially utilized transmission power of nodes besides their relocation. RECRA classifies nodes as critical/non-critical

based on their importance to sustaining connectivity in the network in order to avoid overreacting to non-critical node failures and optimize the recovery from the loss of a critical node. The neighbors of the failed node detect the failure and initiate a recovery process based on their status, proximity, and partially utilized transmission power. To minimize the overhead, RECRA prefers to employ non-critical nodes by moving them to the place of the failed node. If all neighbors are critical, RECRA employs neighbors of the failed node by increasing their transmission power and moving them towards the failed node. The performance of RECRA is validated analytically and through extensive simulations. The simulation results have confirmed the effectiveness of RECRA in sparse and dense topologies and its performance advantage over contemporary schemes found in the literature.

In future, we plan to validate the results of RECRA on a prototype network of mobile robots.

Competing interests

The author(s) declare that they have no competing interests.

Authors' contributions

SS designed and coordinated the study; PV, PS, MA, PDA, BAS and SS conducted the experiments and data analysis. PV, BAS and SS wrote the manuscript and all authors read and approved the final manuscript.

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References

1. I.F. Akyildiz, W. Su, Y. Sankarasubramaniam, E. Cayirci, Wireless sensor networks: a survey. *J. Comput. Netw.* **38**(4), 393–422 (2002)
2. M.B. McMickell, B. Goodwine, L.A. Montestruque, MICAbot: a robotic platform for large-scale distributed robotics, in *Proceedings of the IEEE International Conference on Robotics and Automation (ICRA '03)*, ed. by (Taipei, Taiwan, 2003), pp. 1600–1605
3. K. Dantu, M. Rahimi, H. Shah, S. Babel, A. Dhariwal, G.S. Sukhatme, Robomote: enabling mobility in sensor networks, in *Proceedings of the 4th International Symposium on Information Processing in Sensor Networks (IPSN' 05)*, ed. by (Los Angeles, CA, USA, 2005), pp. 404–409
4. Robot, <http://www.irobot.com>
5. S. Yuta, H. Asama, E. Prassler, T. Tsubouchi, S. Thrun, M. Diel, D. Hahnel, *Scan alignment and 3-D surface modeling with a helicopter platform*, in *Field and Service Robotics*, 24th edn. (Springer Berlin, Heidelberg, 2006), pp. 287–297
6. G. Wang, G. Cao, T.L. Porta, W. Zhang, Sensor relocation in mobile sensor networks, in *Proceedings of the IEEE 24th Annual Joint Conference of the IEEE Computer and Communications Societies (INFOCOM 2005)*, 4th edn. (Miami, 2005), pp. 2302–2312
7. I.F. Akyildiz, I.H. Kasimoglu, Wireless sensor and actor networks: research challenges. *J. Ad Hoc Netw.* **2**(4), 351–367 (2004)
8. CC1000 A unique UHF RF Transceiver. <http://www.chipcon.com>

9. CC2420 2.4 GHz IEEE 802.15.4 ZigBee-ready RF Transceiver. <http://www.chipcon.com>
10. M. Imran, M. Younis, A.M. Said, H. Hasbullah, Localized motion-based connectivity restoration algorithms for wireless sensor and actor networks. *J. Netw. Comput. Appl.* **35**(2), 844–856 (2012)
11. K. Ozaki, K. Watanabe, S. Itaya, N. Hayashibara, T. Enokido, M. Takizawa, A fault-tolerant model for wireless sensor-actor system, in *Proceedings of the 20th International Conference on Advanced Information Networking and Applications (AINA 2006)*, vol. 2 (Vienna, 2006), pp. 303–307
12. R. Ramanathan, R. Rosales-Hain, Topology control of multihop wireless networks using transmit power adjustment, in *Proceedings of the IEEE 19th Annual Joint Conference of the IEEE Computer and Communications Societies (INFOCOM 2000)*, 2nd edn. (Tel Aviv, 2000), pp. 404–413
13. N. Li, J.C. Hou, Topology control in heterogeneous wireless networks: problems and solutions, in *Proceedings of the 23rd Annual Joint Conference of the IEEE Computer and Communications Societies (INFOCOM' 04)* (Hong Kong, 2004), pp. 232–243
14. M. Younis, K. Akkaya, Strategies and techniques for node placement in wireless sensor networks: a survey. *J. Ad Hoc Netw.* **6**(4), 621–655 (2008)
15. P. Basu, J. Redi, Movement control algorithms for realization of fault-tolerant ad hoc robot networks. *J. IEEE Netw.* **18**(4), 36–44 (2004)
16. D. Shantanu, L. Hai, K. Ajith, N. Amiya, I. Stojmenovic, Localized movement control for fault tolerance of mobile robot networks, in *Proceedings of The First IFIP International Conference on Wireless Sensor and Actor Networks (WSAN 2007)* (Albacete, 2007), pp. 1–12
17. A.A. Abbasi, K. Akkaya, M. Younis, A distributed connectivity restoration algorithm in wireless sensor and actor networks, in *Proceedings of the 32nd IEEE Conference on Local Computer Networks (LCN' 07)* (Dublin, Ireland, 2007), pp. 496–503
18. A. Alfadhly, U.A. Baroudi, M. Younis, Least distance movement recovery approach for large scale wireless sensor and actor networks, in *Proceedings of the 7th International Wireless Communications and Mobile Computing Conference (IWCMC'11)* (Istanbul, 2011), pp. 2058–2063
19. M. Younis, S. Lee, S. Gupta, K. Fisher, A localized self-healing algorithm for networks of moveable sensor nodes, in *Proceedings of the IEEE Global Telecommunications Conference (GLOBECOM' 08)* (LA, 2008), pp. 1–5
20. M. Imran, M. Younis, A. Md Said, H. Hasbullah, Volunteer-instigated connectivity restoration algorithm for wireless sensor and actor networks, in *Proceedings of the 2010 IEEE International Conference on Wireless Communications, Networking and Information Security (WCNIS)* (Beijing, 2010), pp. 679–683
21. N. Tamboli, M. Younis, Coverage-aware connectivity restoration in mobile sensor networks. *J. Netw. Comput. Appl.* **33**(4), 363–374 (2010)
22. K. Akkaya, S. Janapala, Maximizing connected coverage via controlled actor relocation in wireless sensor and actor networks. *J. Comput. Netw.* **52**(14), 2779–2796 (2008)
23. M. Duque-Anton, F. Bruyaux, P. Semal, Measuring the survivability of a network: connectivity and rest-connectivity. *Eur. Trans. Telecommun.* **11**(2), 149–159 (2000)
24. D. Goyal, J.J. Caffery, Partitioning avoidance in mobile ad hoc networks using network survivability concepts, in *Proceedings of the Seventh International Symposium on Computers and Communications (ISCC'02)* (Giardini Naxos, 2002), pp. 553–558
25. M. Jorgic, I. Stojmenovic, M. Hauspie, D. Simplot-ryl, Localized algorithms for detection of critical nodes and links for connectivity in ad hoc networks, in *Proceedings of the 3rd Annual IFIP Mediterranean Ad Hoc Networking Workshop (Med-Hoc-Net)* (Bodrum, 2004), pp. 360–371
26. L.H.A. Correia, D.F. Macedo, A.L. dos Santos, A.A.F. Loureiro, J.M.S. Nogueira, Transmission power control techniques for wireless sensor networks. *Comput. Netw.* **51**(17), 4765–4779 (2007)
27. X. Liu, L. Xiao, A. Kreling, Y. Liu, Optimizing overlay topology by reducing cut vertices, in *Proceedings of the 2006 International Workshop on Network and Operating Systems Support for Digital Audio and Video* (Rhode Island, USA, 2006)

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